

1 **Is the surface forcing through sea ice leads transferred**
2 **to the Arctic Ocean interior?**

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6 **Key Points:**

- 7 • A high-resolution simulation allows us to estimate the integrated impact of leads
8 in the Arctic Ocean.
- 9 • Leads make up a quarter of the sea ice cover and buoyancy fluxes through leads
10 are larger than those outside of the leads.
- 11 • Leads account for a quarter of the surface water mass transformation in the Arc-
12 tic Ocean, proportional to their area coverage.

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Abstract

The Arctic sea ice, in particular the ice pack, acts as an insulator between the atmosphere and the ocean. Leads, commonly found in the Arctic, facilitate ocean-atmosphere flux exchanges. Local observations have captured heat fluxes through some leads one order of magnitude larger than those outside of the leads, leading to the speculation that air-sea exchanges through leads contribute significantly to the Arctic Ocean surface buoyancy forcing. Here, we quantify the magnitude and impact on the ocean surface of the leads using SEDNA, a subkilometer pan-Arctic hindcast. Leads account for 22% of the sea ice cover surface, and within them, there is approximately 25% of the total surface water mass transformation. In other words, the water mass transformation in leads is similar to those underneath the surrounding ice-covered oceans. Thus, the present estimate indicates that leads have a small contribution to Arctic Ocean dynamics, contrary to previous hypotheses.

Plain Language Summary

Arctic sea ice acts as a barrier between the air and the ocean, but openings in the ice, called leads, allow for exchanges of heat, salt, moisture, and gases. These leads can significantly increase the amount of heat passing between the ocean and the atmosphere. However, it has been challenging to measure the impact of leads on the ocean because of limited observations and high-resolution models. Using a high-resolution model called SEDNA, we studied the effects of leads across the Arctic. We found that leads cover 22% of the sea ice and explain around 25% of the surface density changes within the ice-covered Arctic. This means the impact of leads on the Arctic Ocean is explained by their area extent in the Arctic. Our main results suggest that leads have a smaller effect on Arctic Ocean dynamics than previously thought.

1 Introduction

The Arctic sea ice regulates Earth's climate by acting as a natural insulator between the atmosphere and the ocean (Wettlaufer et al., 1997; Untersteiner, 1961). A ubiquitous feature of Arctic sea ice is the formation and persistence of leads. Leads occur across the polar regions, both in the marginal ice zone (MIZ; 15–80% sea ice concentration) and the ice pack ($> 80\%$ sea ice concentration). Within the MIZ, leads are primarily formed by the advection of sea ice, while in the ice pack, leads form through the defor-

44 mation of the sea ice and are commonly referred to in the literature as “linear kinematic
45 features” (LKF). The primary forcing of the sea ice advection and deformation is the wind
46 (Linow & Dierking, 2017; Hutter et al., 2018), with a smaller contribution from ocean
47 currents (Willmes et al., 2023). Leads result in openings of the sea ice cover (Rampal
48 et al., 2016; Hutchings et al., 2005; Richter-Menge et al., 2002) with spatial scales of me-
49 ters to kilometers in width and a few kilometers up to hundreds of kilometers in length,
50 and temporal scales ranging from a few hours up to a few days (Linow & Dierking, 2017;
51 Wernecke & Kaleschke, 2015; Tschudi et al., 1998). Thus, they can locally impact the
52 ocean surface forcing in the ice-covered oceans (Lüpkes et al., 2008).

53 Once leads are formed, atmospheric forcing in conjunction with localized upwelling
54 or downwelling occurring in the ocean surface layer can result in important heat fluxes
55 at the ocean surface ($\mathcal{O} \sim 100W/m^2$; Bourgault et al. (2020); Marcq and Weiss (2012);
56 McPhee et al. (2005); Maykut (1986)), instigating localized melting or freezing of sea ice
57 (von Albedyll et al., 2022). For example, Boutin et al. (2023) estimated that in the ice
58 pack 35% of the total sea ice growth occurs within the leads during winter. The Arctic
59 Ocean is a β -ocean, i.e. its stratification is mainly controlled by salinity, which in turn
60 is largely determined by ice-ocean interactions (sea ice growth and melt). Thus, leads
61 experiencing an increase in buoyancy forcing due to brine rejection can induce convec-
62 tion (D. C. Smith & Morison, 1998), weaken the mixed layer stratification, and gener-
63 ate fronts, mixed layer turbulence, and eddies (Reiser et al., 2020; D. C. Smith et al.,
64 2002). Meanwhile, in leads experiencing melting, there will be an increase of the mixed
65 layer stratification and a stabilization of the ocean surface layer. These changes in the
66 buoyancy flux translate into a local transformation of the surface water masses that could
67 be critical for the functioning of the Arctic Ocean circulation (Lenn et al., 2022; Pem-
68 erton et al., 2015; S. D. Smith et al., 1990). Previous studies have shown that instances
69 of leads impact locally the ocean surface and the properties of the Arctic Ocean mixed
70 layer. Nonetheless, the integrated contribution of leads to the large-scale buoyancy forc-
71 ing at the Arctic Ocean surface has not been quantified yet.

72 Here, we present the first estimate of the contribution of the buoyancy forcing and
73 water mass transformation within leads in the Arctic Ocean, using the output of a seven
74 year long pan-Arctic hindcast run at a subkilometer resolution (SEDNA, Talandier and
75 Lique (2023)). Our main objectives are: (1) to assess the impact of leads on the surface

76 buoyancy forcing and (2) to contrast the surface-forced water mass transformation within
77 the leads and those of the ice cover excluding leads.

78 **2 Methods**

79 **2.1 Model description**

80 SEDNA is a $1/60^\circ$ pan-Arctic ocean-sea ice configuration based on the NEMO nu-
81 merical platform (release 4.0.5; Madec et al. (2022)) including the SI3 sea ice component
82 NEMO Sea Ice Working Group (2022). The pan-Arctic domain covers the Bering Strait
83 on the Pacific side and extends southward to 56°N in the Subpolar North Atlantic and
84 to the Baltic Sea entrance. The ocean model solves the primitive equations using finite
85 differences on an Arakawa C-grid. The ocean model incorporates a linear free surface,
86 utilizes a 3rd order flux form scheme for momentum advection, and employs a 4th or-
87 der Flux Corrected Transport (FCT) for tracer advection. Additionally, the sea ice model
88 uses an Elasto-Visco-Plastic (EVP) rheology with a 5-categories sea ice thickness dis-
89 tribution and a landfast ice parameterization to better simulate sea ice behavior above
90 shelves. The NCAR bulk formula from Large and Yeager (2009) is used to calculate air-
91 sea fluxes based on the hourly ERA5 atmospheric data for near-surface variables (Hersbach
92 et al., 2020). The three lateral boundaries are constrained with daily GLORYS12V1 Re-
93 analysis data (Jean-Michel et al., 2021). Monthly freshwater fluxes from river discharges
94 are combined with Greenland land ice melt data (Hu et al., 2019). The simulation ex-
95 tends over seven years (2009 to 2015) and starts from rest, utilizing initial conditions based
96 on the World Ocean Atlas 2009 temperature/salinity and mean January 2009 ice state
97 from PIOMAS re-analysis (Locarnini et al., 2010; Zhang & Rothrock, 2003). A weak restor-
98 ing toward the World Ocean Atlas climatological sea surface salinity is applied, but only
99 over the ice-free regions. The results presented hereafter correspond to the daily mean
100 outputs during 2014 to allow the ocean surface to equilibrate after 6 years of spin-up.

101 **2.2 Identification of leads**

102 The leads from SEDNA are identified by using the algorithm proposed by Hutter
103 et al. (2019) applied to the daily mean sea ice outputs. This method uses the total de-

104 formation rate defined as:

$$T_d = \left[\underbrace{\left(\frac{\partial u}{\partial x} + \frac{\partial v}{\partial y} \right)^2}_{Divergence} + \underbrace{\left(\frac{\partial u}{\partial x} - \frac{\partial v}{\partial y} \right)^2 + \left(\frac{\partial u}{\partial y} + \frac{\partial v}{\partial x} \right)^2}_{Shear} \right]^{1/2}, \quad (1)$$

105 where u and v are the sea ice velocities. The deformation rate varies spatially depend-
 106 ing on the background deformation and the sea ice properties (Reiser et al., 2020), and
 107 the largest deformation rates occur along the boundaries of ice floes where the leads are
 108 found. Therefore, to identify the leads, the local maxima are identified using a difference
 109 of Gaussian filter (DoG filter) of the deformation rate field. After this, a binary map is
 110 created with the mask of the identified lead. Note that after this step, Hutter et al. (2019)
 111 reduces the width of the leads to a line, and then tracks the leads over time. Since our
 112 study focuses on ocean-atmosphere interactions, we only use the identification algorithm
 113 to produce a mask of leads during 2014 for the Arctic, but we retain the original lead
 114 width computed from the deformation rate. Finally, the lead mask is then used to quan-
 115 tify and contrast the impact of leads compared to the ice-covered Arctic, the ice-pack
 116 (ice concentration $> 80\%$), and the MIZ (15 - 80% ice concentration).

117 2.3 Buoyancy flux and water mass transformation

118 Heat fluxes (from shortwave radiation, longwave radiation, latent heat, and sen-
 119 sible heat) and freshwater fluxes (from evaporation, precipitation, runoff, and sea ice melt/growth)
 120 at the ocean surface result in a buoyancy forcing capable of changing the density of the
 121 ocean surface. Here, the buoyancy flux is computed as:

$$B_f = \underbrace{\frac{g\alpha Q_{net}}{\rho_0 C_p}}_{Heat\ Contribution} - \underbrace{\frac{g\beta F_{net} SSS}{\rho_0}}_{Freshwater\ Contribution}, \quad (2)$$

122 where Q_{net} the net heat flux at the sea surface, α the thermal expansion coefficient,
 123 SSS the sea surface salinity, F_{net} the net freshwater flux, β the haline contraction co-
 124 efficient, $g = 9.81m/s^2$, and the ocean density $\rho_0 = 1026kg/m^3$. Note that using this
 125 sign convention, a positive buoyancy forcing decreases the surface density making the
 126 ocean surface more buoyant (stable) and a negative buoyancy forcing increases the sur-
 127 face density. The area-weighted buoyancy ($\langle \rangle$) fluxes are computed following:

$$\langle B_f(t) \rangle = \frac{\sum_x \sum_y (B_f(t, x, y) * Area(x, y) * Mask_R(t, x, y))}{\sum_x \sum_y Area(x, y) * Mask(t, x, y)}, \quad (3)$$

128 where the *Area* is the area of the grid cells, and the $Mask_r$ is the mask of the MIZ, the
 129 Pack, the MIZ leads, the ice pack leads, and the ice cover that varies in space and time.
 130 Note that the divisor *Mask* corresponds to the full ice-covered area (including leads).

131 Walin (1982) and Nurser et al. (1999) proposed a surface-forced water mass trans-
 132 formation framework that provides information on the transport of water through the
 133 surface density contours due to surface buoyancy forcing that results in the formation
 134 of lighter or denser water at the surface. Thus, the WMT relates the surface buoyancy
 135 forcing and the surface density field to the properties of the ocean interior. A negative
 136 transformation rate corresponds to a water mass becoming lighter, and a positive trans-
 137 formation rate corresponds to a water mass becoming denser. The water mass transfor-
 138 mation (WMT) is computed as:

$$\Omega(\sigma_k) = -\frac{1}{\sigma_{k+1} - \sigma_k} \int \int_A B_f dA, \quad (4)$$

139 where σ is potential density and the indices k represents the density bin number.
 140 The computation was performed with 172 density bins of $0.1kg/m^3$ within the density
 141 range of $15.15kg/m^3 - 32.25kg/m^3$. The transformation rate Ω is spatially decomposed
 142 into different sea ice masks that are the ice-covered, the pack ice, the MIZ, and the leads
 143 in each of these regions.

144 **3 Buoyancy forcing in leads**

145 Leads can be easily visually identified when examining high resolution satellite ob-
 146 servations (Fig. 1f), yet it remains challenging to capture them in state-of-the-art mod-
 147 els that often lack the resolution and/or the sea ice dynamics to simulate these features
 148 (Bouchat et al., 2022; Wang et al., 2016). Here we analyze daily outputs from the SEDNA
 149 ocean-sea ice model, which has an average Arctic resolution of $\sim 800m$, sufficient to cap-
 150 ture the observed lead width distribution ($> 1km$; Wernecke and Kaleschke (2015)). We
 151 start our analysis by detecting the leads in the model sea ice fields, using the detection
 152 algorithm of Hutter et al. (2019) based on the sea ice deformation field (see Eq.1). Fig-
 153 ure 1a and c shows a snapshot of the sea ice deformation for the 21st of April 2014, from
 154 which we retrieve a mask of the leads (Fig. 1e). Within the leads, the sea ice concen-
 155 tration can remain higher than 90% (Fig. 1d); this is a consequence of the continuous

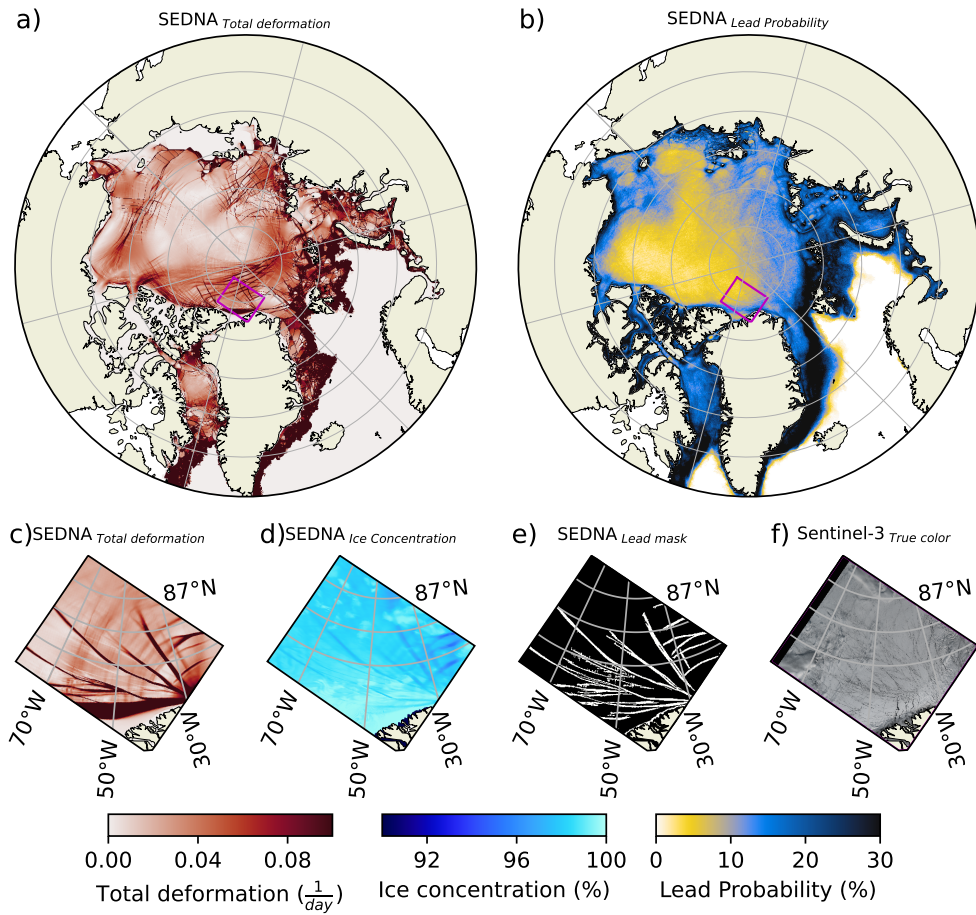


Figure 1. a) Snapshot of total deformation for the 21st of April 2014 from SEDNA. b) Mean lead probability for daily snapshots of leads identified between 2011 and 2015 from SEDNA. Zoom of the magenta box of panel a for the c) total deformation to identify leads, d) ice concentration, and f) identified mask of leads. e) True color image from Sentinel 3 on the 12th of April 2022 for the magenta area in panel a. Note that the true color image is not associated with any colorbar.

156 nature of the sea ice model and the daily averaged output that smooths the presence of
 157 leads in the fields.

158 The SEDNA climatology of the lead probability (Fig. S1a), i.e. the likelihood of
 159 leads occurring for any given day is $\sim 19\%$ in the ice-covered Arctic, comparable to pre-
 160 vious observations (Wernecke & Kaleschke, 2015; Willmes et al., 2023). Although ob-
 161 servations use different identification methods compared to the one implemented in SEDNA
 162 (Hutter et al., 2019), there is resemblance between SEDNA and observations (Fig. S1b),
 163 with similar lead probability over the shelves, along bathymetric features, and in regions
 164 where the ocean surface kinetic energy tends to be large (Fig. S1c). Yet, SEDNA un-
 165 derestimates the observed lead probability in the Beaufort Sea. Regardless, the similar-
 166 ities between the observations and the model lead probability give us confidence to quan-
 167 tify for the first time the impact of leads on the surface buoyancy of the Arctic Ocean.

168 Figure 2a and b show two snapshots of the buoyancy flux at the ocean surface. A
 169 negative buoyancy flux results from a loss of heat to the atmosphere, sea ice refreezing,
 170 and brine rejection, while a positive buoyancy flux is associated with sea ice melting, fresh-
 171 ening, and warming of the Arctic surface. Inside the ice-covered region, the fluxes are
 172 spatially homogeneous, except in the MIZ and within the leads, where the buoyancy fluxes
 173 are larger. The average buoyancy flux across the ice-covered Arctic (15% - 100% ice con-
 174 centration) within the leads is $-2.2 \times 10^{-8} m^2/s^3$ and $7.2 \times 10^{-8} m^2/s^3$ for each date,
 175 compared to the $-2.4 \times 10^{-8} m^2/s^3$ and $5.7 \times 10^{-8} m^2/s^3$ excluding the leads. In other
 176 words, on the 1st of January, the total flux within the leads and outside of the leads are
 177 comparable. In contrast, on the 25th of June, the total buoyancy flux in the leads is up
 178 to 30% larger than those over the ice cover region excluding leads.

179 Examining a transect in the Canadian Basin on the 1st of January 2014, we de-
 180 tect two leads (vertical blue bars) associated with a significant decrease in sea ice thick-
 181 ness of $\sim 25cm$, suggesting an opening of the sea ice cover (Fig.2 c). This transect ex-
 182 hibits small heat flux at the ocean surface and within the identified leads (Fig.2 e). This
 183 is because the ocean surface is at freezing point and the heat exchanged with the atmo-
 184 sphere is used to form sea ice rather than cooling the ocean. The freshwater flux (Fig.2g)
 185 exhibits two prominent positive peaks located within the identified leads. Both peaks
 186 are twice the magnitude of the fluxes outside the leads, highlighting enhanced sea ice growth
 187 and brine rejection within the leads. The total buoyancy flux is negative for this snap-

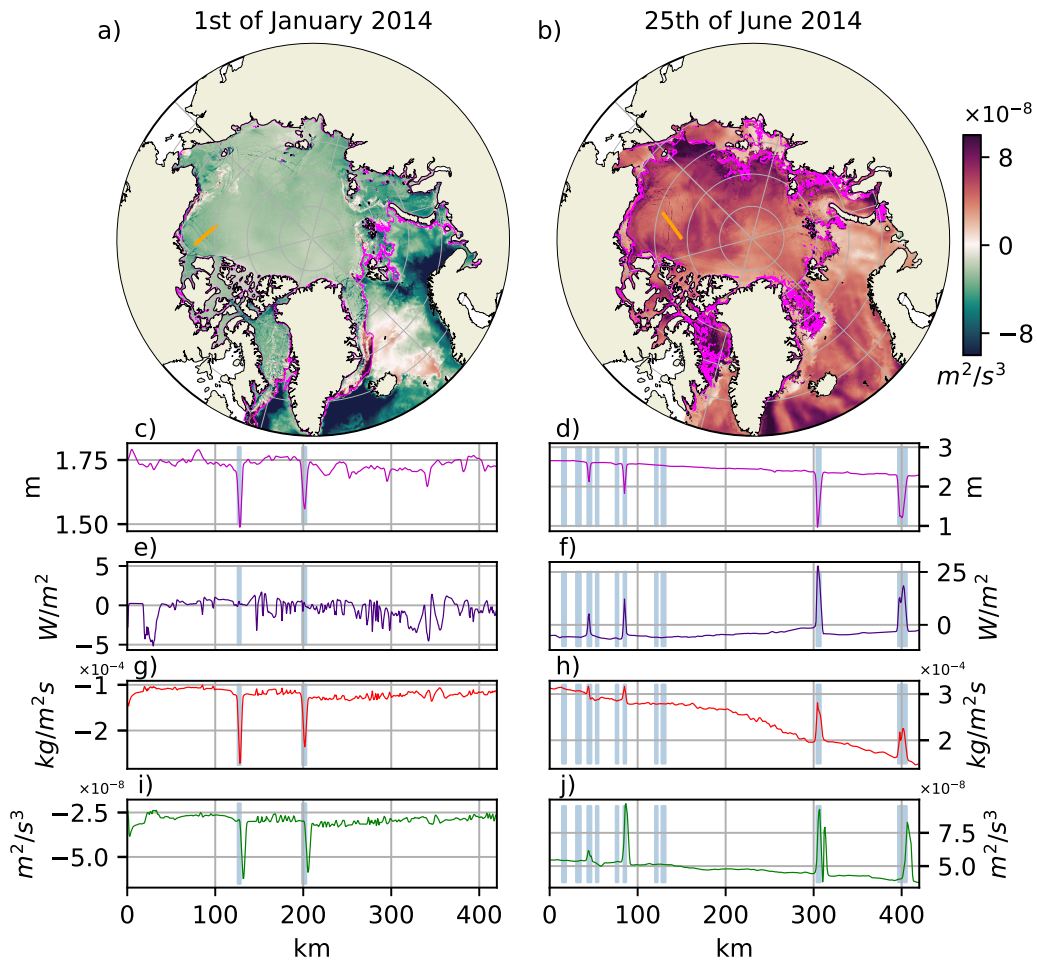


Figure 2. Maps of buoyancy flux on (a) the 1st of January 2014 and (b) the 25th of June 2014. Transect on the 1st of January 2014 of c) ice thickness, e) downward heat flux, g) freshwater flux (positive upward), and i) buoyancy flux (heat flux minus freshwater flux). Transect on the 25th of June 2014 of d) ice thickness, f) downward heat flux, h) freshwater flux (positive upward), and j) buoyancy flux. Note that positive buoyancy fluxes correspond to buoyancy loss, while negative buoyancy fluxes correspond to buoyancy gain by the ocean. The transect location and orientation changes between the winter and summer dates. Magenta contour in panels a and b corresponds to the 15% concentration contour of sea ice. Vertical blue bars show the mask of leads in the transect.

188 shot, with the largest buoyancy loss occurring within the leads (Fig.2 i). On the 25th
 189 of June (Fig.2d), we detect several leads, but only some show a decrease of the sea ice
 190 thickness of more than 1m (Fig.2 h). These specific leads also experience an increase in
 191 heat fluxes, reaching up to $\sim 25W/m^2$ (Fig.2f) and a local decrease of the freshwater
 192 flux, indicating a freshening of the surface due to sea ice melt. This change is captured
 193 by the net positive buoyancy flux, i.e. a decrease in the density at the surface (Fig.2j).
 194 Of all the identified leads, some exhibit significant fluxes at the ocean surface, while oth-
 195 ers have little to no imprint in the thickness and fluxes. For some, this is explained by
 196 the timing of the sea ice break up, as subsequent snapshots show an increase of fluxes
 197 at the location of these identified leads (not shown). Alternatively, previous studies have
 198 shown that in leads generated through shear, there is a limited buoyancy exchange with
 199 the atmosphere, but may still enhance fluxes in the ocean due to isopycnal upwelling or
 200 downwelling forced by the atmosphere-ocean stress within the leads (Bourgault et al.,
 201 2020).

202 Similar to observations, we identify buoyancy fluxes that are larger within the leads
 203 than outside of them for these summer and winter days. In the next section, we explore
 204 the spatially averaged contribution of leads to the surface buoyancy fluxes and the in-
 205 tegrated influence on the water mass transformation across the Arctic Basin during 2014.

206 **4 Integrated contribution of leads to the Arctic buoyancy**

207 The following analysis is conducted separately for four distinct masks represent-
 208 ing regions with leads and regions excluding leads for the MIZ and the ice pack. On av-
 209 erage over 2014, these masks account for 6% for the MIZ, 72% for the ice pack, 6% for
 210 the leads in the MIZ, and 16% for the leads in the ice pack. Thus the part covered by
 211 leads is on average $\sim 22\%$ of the total sea ice-covered Arctic, consistent with previous
 212 model assessments (Wang et al., 2016). The extent of each mask exhibits a seasonal cy-
 213 cle (Fig. 3a). From winter (December, January, and February) to summer (June, July,
 214 and August), the ice pack extent decreases from approximately 80% to 60%, while the
 215 extent of the MIZ increases from 2% to approximately 13%. On one hand, the percent-
 216 age of the ice pack leads extent peaks in winter at 16% and decreases to a minimum ex-
 217 tent of 13% in summer. On the other hand, the extent of the MIZ leads covers 2% in
 218 winter and around 13% in summer, the same as the MIZ extent.

219 The contribution of the total buoyancy flux occurring within the leads and ice-covered
 220 regions is quantified by weighting the buoyancy flux of each of the sea ice cover masks
 221 by the total ice-covered area (Fig. 3b; Eq.3; note that the size of each mask evolves in
 222 time). The mean buoyancy flux under the ice pack in winter (December-February) and
 223 summer (June-August) is $-1.8 \times 10^{-8} m^2/s^3$ and $2.2 \times 10^{-8} m^2/s^3$, respectively. The
 224 buoyancy flux underneath the ice pack changes signs during the seasonal cycle, suggest-
 225 ing a shift from ice formation in winter to ice melting in summer. Note that in the ice
 226 pack, the haline component of the buoyancy flux determines the buoyancy flux, while
 227 the heat contribution is negligible (see Fig.S2 for a quantification of the different con-
 228 tributions). Meanwhile, the mean buoyancy flux in the MIZ is negligible in winter and
 229 increases to $0.9 \times 10^{-8} m^2/s^3$ in summer. The MIZ is predominantly associated with a
 230 positive buoyancy flux suggesting preferential melting in this region throughout the year
 231 (except for autumn). There, the haline component also determines the buoyancy flux
 232 (See Fig.S2). In both, the MIZ leads and the ice pack leads, the buoyancy flux follows
 233 the same seasonality of the fluxes outside of the leads. Winter values are $-0.3 \times 10^{-8} m^2/s^3$
 234 and $-0.01 \times 10^{-8} m^2/s^3$, while summer values reach $0.6 \times 10^{-8} m^2/s^3$ and $0.9 \times 10^{-8} m^2/s^3$
 235 for the contribution of the leads in the ice pack and the MIZ, respectively. Overall, leads
 236 explain between 20% to 45% of the ice-covered buoyancy fluxes (red dashed line in Fig
 237 3b), more than the area extent of all leads that ranges from 18% in winter to 26% in sum-
 238 mer. Therefore, fluxes are larger within the leads relative to their area extent.

239 The role and impact leads have on the surface water masses of the Arctic can be
 240 quantified by using the surface water mass transformation framework (Eq. 4; Walin (1982)).
 241 Figure 4a shows the averaged WMT during 2014 for the total ice-covered Arctic, the ice
 242 pack, the MIZ, and the parts of these regions covered by leads. The definitions of the
 243 water masses depicted in figure 4a follow roughly those proposed by Pemberton et al.
 244 (2015) and Lansard et al. (2012). The yearly mean WMT in the ice pack is character-
 245 ized by a broad region of strong positive transformation (losing buoyancy) correspond-
 246 ing approximately to the halocline polar surface water (PSW_H ; $24.4 kg/m^3 < \rho < 27.4 kg/m^3$)
 247 reaching a maximum of $\sim 1.8 Sv$, consistent with previous estimates of the surface Arc-
 248 tic WMT (Pemberton et al., 2015). The yearly mean WMT in the ice pack leads also
 249 features a positive peak within the PSW_H density class reaching up to $0.4 Sv$. On av-
 250 erage during 2014, the WMT in the ice pack leads is $\sim 20\%$ of the WMT in the ice pack
 251 (without leads). The mean MIZ WMT is mostly negative within the density range of

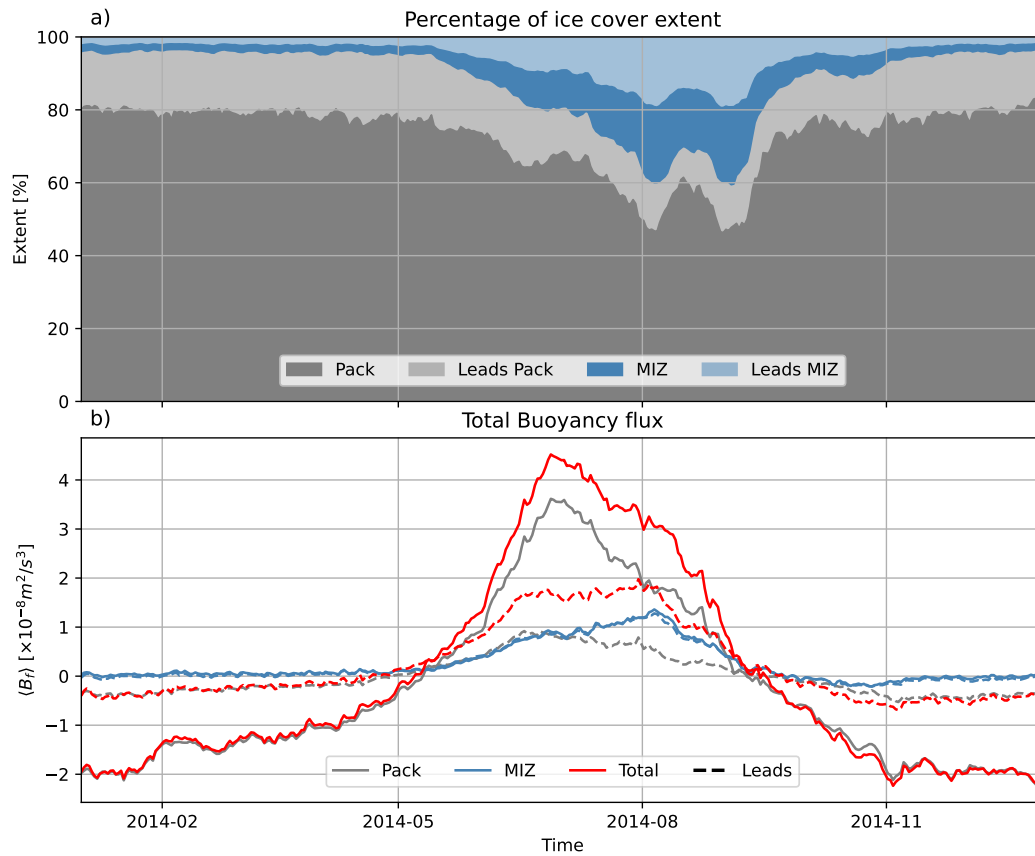


Figure 3. Time series during 2014 of a) the percentage of ice cover extent and b) the total buoyancy fluxes for the total ice-covered, the ice pack, the MIZ, and the leads in the ice pack and the MIZ. The MIZ is defined between ice concentration of 15-80%, the ice pack is defined where ice concentration $> 80\%$, and the total ice-covered is defined as the sum of the MIZ and ice pack.

252 $20\text{kg}/\text{m}^3 < \rho < 27.5\text{kg}/\text{m}^3$ (gaining buoyancy), reaching a minimum of -0.4Sv around
 253 $\rho = 27\text{kg}/\text{m}^3$. The MIZ leads WMT is also negative within the same range of densi-
 254 ties, with a minimum of -0.4Sv . Thus, the MIZ leads WMT is comparable to those in
 255 the MIZ excluding leads. Hovmöller diagrams of the daily WMT estimates over 2014 for
 256 the four masks are also depicted in figure 4. The mean WMT underneath the ice pack
 257 varies from 1.1 Sv in winter to -1.0 Sv in summer, while the WMT in the leads in the
 258 ice pack is 0.2 Sv in winter and -0.26 Sv in summer. In other words, the WMT in the
 259 ice pack leads ranges from 20 to 25% of the WMT occurring in the pack (Fig. 4d). It
 260 is interesting to note that the patterns across the density classes of the WMT for the leads
 261 and the surrounding regions excluding leads are similar (Figs. 4b and c). This suggests
 262 that there is no specific water mass that is transformed preferentially with the leads. Rather,
 263 our results suggest that the water masses transformed within the leads and outside of
 264 them have similar properties. The WMT in MIZ leads and those in the MIZ have com-
 265 parable magnitudes over the year (Fig. 4 g). Yet, similarly to the leads in the ice pack,
 266 there is no evidence that leads in the MIZ could transform preferentially specific water
 267 masses compared to the remaining MIZ. Adding the contribution of the WMT in all the
 268 leads in the Arctic (those in the MIZ and the ice pack), we estimate $\sim 25\%$ of the WMT
 269 happens in leads, which is comparable to the surface covered by these leads in the Arc-
 270 tic ($\sim 22\%$).

271 5 Conclusions

272 The present study assesses the impact of leads on the ocean surface by diagnos-
 273 ing the buoyancy fluxes using a very high-resolution hindcast (SEDNA). SEDNA cap-
 274 tures the intermittency of the leads and the lead properties. Leads cover $\sim 22\%$ of the
 275 sea ice-covered Arctic, and in the ice pack, leads can contribute more buoyancy flux com-
 276 pared to the fluxes outside the leads. Despite these larger buoyancy fluxes, leads only
 277 account for approximately $\sim 25\%$ of the total surface water mass transformation in the
 278 ice-covered Arctic, which is roughly equivalent to their surface coverage. Thus, contrary
 279 to previous hypotheses based largely on local and intermittent observations (e.g. McPhee
 280 et al. (2005); Morison and McPhee (1998)), our findings suggest that leads have a small
 281 contribution to the transformation of surface water masses in the Arctic Ocean. While
 282 their presence can induce localized changes in the density of surface waters due to brine
 283 rejection and ice melting, the transformation of water masses in the leads has the same

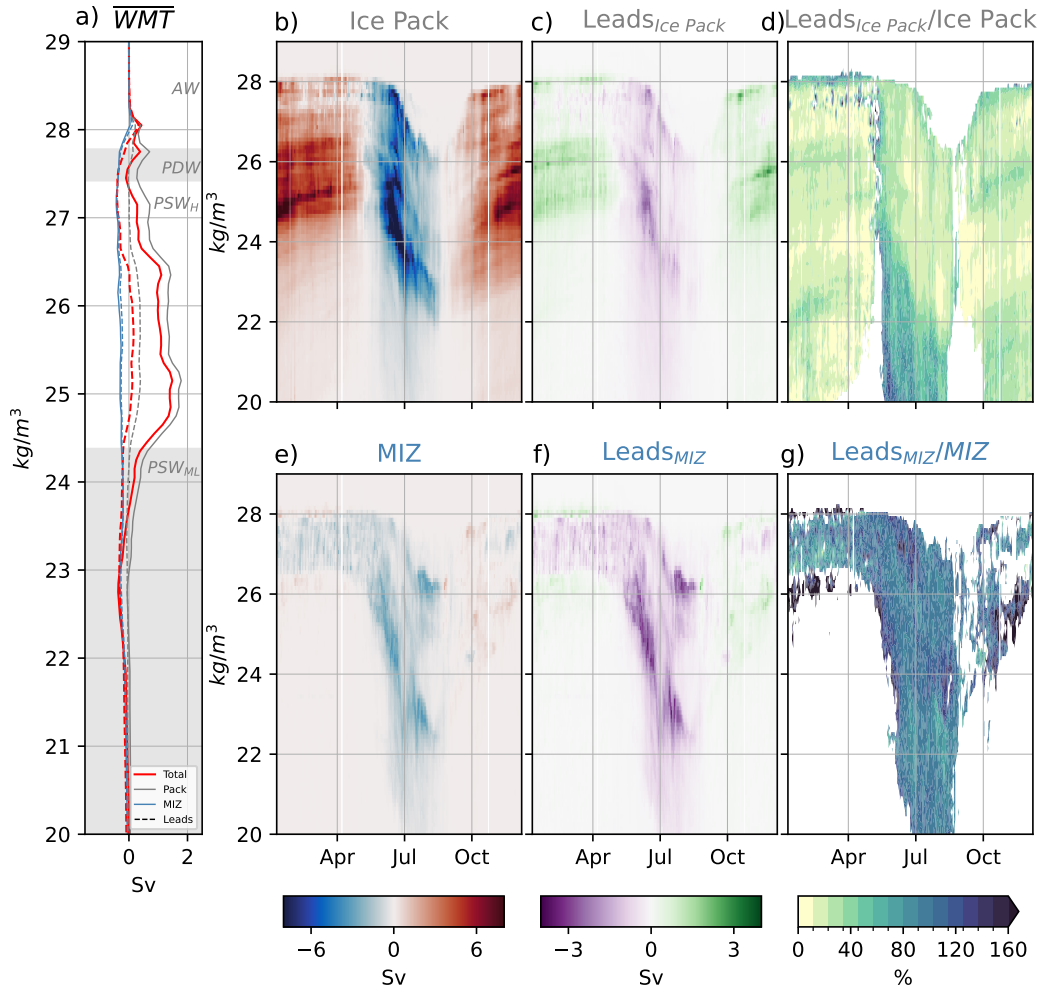


Figure 4. a) Mean water mass transformation over 2014 for the total ice-covered, ice pack, MIZ, and leads in each region. Hovmöller diagram of the water mass transformation (Sv) for b) the ice pack, c) the ice pack leads, and d) the ratio of water mass transformation between the leads and the ice pack region. Panels e), f), and g) follow the same structure as panels b), c), and d), but for the MIZ, MIZ leads, and the ratio between them. Shaded gray and white areas in panel a) represent approximately the water masses of the Arctic Ocean according to Pemberton et al. (2015) and Lansard et al. (2012): the mixed layer polar surface water (PSW_{ML} ; $\rho < 24.4kg/m^3$), the halocline polar surface water (PSW_H ; $24.4kg/m^3 < \rho < 27.4kg/m^3$), the polar deep water (PDW ; $27.4kg/m^3 < \rho < 27.8kg/m^3$), and the Atlantic water (AW ; $27.8kg/m^3 < \rho$).

284 sign and magnitude as those underneath the surrounding ice-covered regions. Thus, no
 285 evidence was found that leads impact in a distinct way the Arctic Ocean water masses
 286 in SEDNA compared to outside the leads in the ice-covered ocean.

287 Here we solely focus on the surface forced WMT, but we acknowledge that leads
288 could further impact properties at the surface and interior of the ocean by enhancing mix-
289 ing within leads, inducing upwelling of isopycnals through Ekman pumping (McPhee et
290 al., 2005), modifying entrainment in the mixed layer within the leads, providing a source
291 of potential energy, and generating mixed layer instabilities due to isopycnal tilting in
292 the vicinity of leads. The use of a continuous sea ice model that results in leads with high
293 ice thickness and concentrations likely underestimates the exchanges between the ocean
294 and the atmosphere. Therefore, future studies using floe-resolving models that allow more
295 “realistic leads” with direct exchanges between the ocean and the atmosphere may re-
296 sult in larger fluxes through the leads increasing their potential impact in the water mass
297 transformation and ocean mixed layer.

298 Leads have an important local effect in the atmosphere and the sea ice fluxes (Marcq
299 & Weiss, 2012; Lüpkes et al., 2008; Wettlaufer et al., 1997; S. D. Smith et al., 1990), but
300 a limited imprint on the ocean surface and interior properties. Our results suggest that
301 the presence of leads is not the first order mechanism controlling the water mass trans-
302 formation forced at the surface, thus leads have a minor impact on the stratification and
303 surface layer dynamics of the Arctic Ocean. Therefore, resolving and/or parameteriz-
304 ing leads in climate models are likely to only marginally improve the misrepresentation
305 of the stratification and water masses found in Arctic climate models (Wang et al., 2024;
306 Ilıcak et al., 2016). Finally, the probability of leads occurring in the ice pack has been
307 projected to increase as the Arctic sea ice transitions towards thinner, younger, and more
308 mobile sea ice (Boutin et al., 2023; Intergovernmental Panel on Climate Change (IPCC),
309 2022). Our analysis is only performed over one year and the future impact of more abun-
310 dant leads is not represented in our simulation. Thus, dedicated studies are required to
311 better estimate the surface buoyancy forcing within leads in the context of an evolving
312 Arctic Ocean.

313 **Open Research Section**

314 The SEDNA model configuration is described and publicly available via Talandier
315 and Lique (2023). All analyses and figures in this manuscript can be reproduced using
316 the Jupyter notebooks and instructions provided in the Zenodo archive [Leads_Arctic_ocean](#),
317 Martínez-Moreno et al. (2024).

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